

Rank correlation between Two Interval Valued Intuitionistic fuzzy Sets

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Abstract:

In this paper we investigated the correlation between two interval valued intuitionistic fuzzy sets defined on the same universal support. It is shown that Spearman's rank correlation coefficient can be applied in the members of the supports are ranked according to the interval valued intuitionistic fuzzy membership and non-membership values of each set. Next a membership and non-membership value-based interval valued intuitionistic fuzzy rank correlation measure is proposed and its example are derived.

Keywords:

Correlation, Random variable, Spearman's Rank correlation, Interval valued Intuitionistic fuzzy sets .

Introduction:

Interval valued Intuitionistic Fuzzy set theory is a powerful tool to model imprecise and vague situations where exact analysis is either difficult or impossible. Central to the theory is the concept of Interval valued intuitionistic fuzzy set, which is generalization of the ordinary (crisp) set that captures gradual transition from belongingness to the set .

It is often theoretically interesting and practically necessary to see how different properties of crisp set can be generalized in Interval valued intuitionistic fuzzy situations. Such studies led to enormous interesting result.

In statistics and in engineering sciences a concept called rank correlation is often used. By rank correlation analysis the joint relationship of two variables can be examined with the aid of a measure of interdependence of the two variables. Typical subjects for rank correlation analysis are the interdependence (association) between height of father and height of (adult) son, between score in mathematics and score in statistics, etc. In all these cases the value taken by the variables is objective and the value obtained is the outcome of a purely random event.

If $(x_i, y_i; i=1,2,\dots,n)$ is the data set, then the rank correlation between x and y is defined as $\rho = 1 - \frac{6d^2}{n(n^2-1)}$ where $d_i = R_1 - R_2$ is the rank between x and y . But in reality there are many situations where instead of the measured values of the two variables of some objects or entities the ranks of the objects according to the two variables or according to two different qualitative characteristics is available. In this case rank correlation is used to correlate the two sets of available data. Such a situation arises when we have a qualitative rather than quantitative knowledge of the samples which can be ranked according to quality. Consider, for example, the quality of handwriting of individuals which has no obvious way to quantify. A typical rank correlation measure is due to Spearman [2]. But an interesting case arises if instead of ranks some subjective grading of the elements is available (the grading may be continuous or discrete). Then a interval valued intuitionistic fuzzy set can be formed based on such grading and the relation

between two such interval valued intuitionistic fuzzy sets can be obtained by comparing the respective membership and non-membership values in the sets.

INTERVAL-VALUED INTUITIONISTIC FUZZY SET

An interval-valued intuitionistic fuzzy set (IVIFS) A in X , $X \neq \emptyset$ and $\text{card}(X)=n$, is an object having the form: $A = \{ \langle x, \mu_A(x), \gamma_A(x) \rangle : x \in X \}$ Where $\mu_A : X \rightarrow [0,1], \gamma_A : X \rightarrow [0,1]$ with the condition $\sup \mu_A(x) + \sup \gamma_A(x) \leq 1$ for any $x \in X$. The intervals $\mu_A(x)$ and $\gamma_A(x)$ denote, respectively, the degree of belongingness and the degree of non-belongingness of the element x to A . We denote by $\text{IVIFS}(X)$ the set of all IVIFSs in X . Then for each $x \in X$, $\mu_A(x)$ and $\gamma_A(x)$ are closed intervals and their lower and upper end points are denoted by $\mu_{AL}(x), \mu_{Au}(x), \gamma_{AL}(x), \gamma_{Au}(x)$ respectively, and thus we can replace with $A = \{ \langle x, [\mu_{AL}(x), \mu_{Au}(x)], [\gamma_{AL}(x), \gamma_{Au}(x)] \rangle : x \in X \}$ where $0 \leq \mu_{Au}(x) + \gamma_{Au}(x) \leq 1$ for any $x \in X$. For each $A \in \text{IVIFS}(X)$ we call $\pi_A(x) = [1 - \mu_{Au}(x) - \gamma_{Au}(x), 1 - \mu_{AL}(x) - \gamma_{AL}(x)]$ an intuitionistic fuzzy interval of X in A . Its lower and upper points are $\pi_{AL}(x) = 1 - \mu_{Au}(x) - \gamma_{Au}(x)$ and $\pi_{Au}(x) = 1 - \mu_{AL}(x) - \gamma_{AL}(x)$, respectively.

The following expressions are defined for $A, B \in \text{IVIFS}(X)$;

$$A \leq B \Leftrightarrow \mu_A(x) \leq \mu_B(x) \text{ and } \gamma_A(x) \geq \gamma_B(x) \text{ for all } x \in X;$$

$$A = B \Leftrightarrow \mu_A(x) = \mu_B(x) \text{ and } \gamma_A(x) = \gamma_B(x) \text{ for all } x \in X;$$

$$A^c \Leftrightarrow \{ \langle x, \mu_A(x), \gamma_A(x) \rangle : x \in X \}.$$

$$A < B \Rightarrow \mu_A(x) \leq \mu_B(x) \text{ and } \gamma_A(x) \leq \gamma_B(x) \text{ for all } x \in X.$$

The operational relation and defined for $A, B \in \text{IVIFS}(X)$ and $\lambda > 0$;

$$A \cap B = \{ \langle \min \mu_{AL}(x), \mu_{BL}(x), \min \mu_{Au}(x), \mu_{Bu}(x), [\max(\gamma_{AL}(x), \gamma_{BL}(x)), \max(\gamma_{Au}(x), \gamma_{Bu}(x))] \rangle \};$$

$$A \cup B = \{ \langle \max(\mu_{AL}(x), \mu_{BL}(x)), \max(\mu_{Au}(x), \mu_{Bu}(x)), [\min(\gamma_{AL}(x), \gamma_{BL}(x)), \min(\gamma_{Au}(x), \gamma_{Bu}(x))] \rangle \};$$

$$A + B = \{ \langle [\mu_{AL}(x) + \mu_{BL}(x) - \mu_{AL}(x)\mu_{BL}(x), \mu_{Au}(x) + \mu_{Bu}(x) - \mu_{Au}(x)\mu_{Bu}(x)], [\gamma_{AL}(x)\gamma_{BL}(x), \gamma_{Au}(x)\gamma_{Bu}(x)] \rangle \};$$

$$AB = \{ \langle \mu_{AL}(x)\mu_{BL}(x), \mu_{Au}(x)\mu_{Bu}(x) \rangle \};$$

$$[\gamma_{AL}(x) + \gamma_{BL}(x) - \gamma_{AL}(x)\gamma_{BL}(x), \gamma_{Au}(x) + \gamma_{Bu}(x) - \gamma_{Au}(x)\gamma_{Bu}(x)] \};$$

$$\lambda A = \{ \langle [1 - (1 - \mu_{AL}(x))^\lambda, 1 - (1 - \mu_{Au}(x))^\lambda], [\gamma_{AL}(x)^\lambda, \gamma_{Au}(x)^\lambda] \rangle \};$$

$$A^\lambda = \{ \langle \mu_{AL}(x)^\lambda, \mu_{Au}(x)^\lambda, [1 - (1 - \gamma_{AL}(x))^\lambda, 1 - (1 - \gamma_{Au}(x))^\lambda] \rangle \}.$$

OPERATORS OVER INTERVAL VALUED IFS

The interval valued intuitionistic fuzzy sets (IVIFS_s) were defined as extension as of the intuitionistic fuzzy set (IFS_s). An IVIFS A over E is an object having the form: $A = \{ \langle x, M_A(x), N_A(x) \rangle / x \in E \}$, where $M_A(x) \subset [0,1]$ and $N_A(x) \subset [0,1]$ are intervals and for every $x \in E$, $\sup M_A(x) + \sup N_A(x) \leq 1$. Obviously, when each of the intervals $M_A(x)$ and $N_A(x)$ contains exactly one element each of them i.e., when for every $x \in E: \mu_A(x) = \inf M_A(x) = \sup M_A(x)$,

$\gamma_A(x) = \inf N_A(x) = \sup N_A(x)$, the given IVIFS is transformed to an ordinary IFS. On the other hand, when $N_A(x) = \emptyset$ for every $x \in E$, the given IVIFS is transformed to an ordinary interval valued fuzzy set.

We note that the transformation formulas between an IVIFS and an interval valued fuzzy set are the following:

- a) When the IVIFS A is given, then we can construct, corresponding to it, an interval valued fuzzy set B for which $M_B(x) = [\mu_A(x), 1 - \gamma_A(x)]$;

- a) When the interval valued fuzzy set B is given, we can construct, corresponding to it, an IVIFS A for which

$$\mu_A(x) = \inf M_B(x), \gamma_A(x) = 1 - \sup M_B(x).$$

Here we shall present different operators over the IVIFS_s. They are analogous to the operators, defined over the IFS and on the other hand some of them can also be transformed over the interval valued fuzzy set. Some of the operators are specific and they can be applied only over the IVIFS_s. Situation is analogous to the situation in the IFS theory. For example, operators \square and \diamond transform a given IFS to ordinary sets and we have that.

$$\square A \subset A \subset \diamond A$$

For every IFS A, while for every ordinary fuzzy set

$$\square A = A = \diamond A$$

i.e., the two operators are senseless in the frames of the fuzzy set theory.

For every two IVIFS_S A and B the following relation and operations are valid

$$A \subset_{\square, \inf} B \text{ iff } (\forall x \in E) (\inf M_A(x) \leq \inf M_B(x)),$$

$$A \subset_{\square, \sup} B \text{ iff } (\forall x \in E) (\sup M_A(x) \leq \sup M_B(x)),$$

$$A \subset_{\square, \inf} B \text{ iff } (\forall x \in E) (\inf N_A(x) \geq \inf N_B(x)),$$

$$A \subset_{\square, \sup} B \text{ iff } (\forall x \in E) (\sup N_A(x) \geq \sup N_B(x)),$$

$$A \subset_{\square} B \text{ iff } A \subset_{\square, \inf} B \ \& \ A \subset_{\square, \sup} B,$$

$$A \subset_{\square} B \text{ iff } A \subset_{\square, \inf} B \ \& \ A \subset_{\square, \sup} B,$$

$$A \subset B \text{ iff } A \subset_{\square} B \ \& \ B \subset_{\square} A,$$

$$A = B \text{ iff } A \subset B \ \& \ B \subset A$$

$$A = \{ \langle x, N_A(x), M_A(x) \rangle / x \in E \}$$

$$A \cap B = \{ \langle x, [\min(\inf M_A(x), \inf M_B(x)), \min(\sup M_A(x), \sup M_B(x))], [\max(\inf N_A(x), \inf N_B(x), \max(\sup N_A(x), \sup N_B(x)))] \rangle / x \in X \},$$

$$A \cap B = \{ \langle x, [\max(\inf M_A(x), \inf M_B(x)), \min(\sup M_A(x), \sup M_B(x))], [\min(\inf N_A(x), \inf N_B(x), \min(\sup N_A(x), \sup N_B(x)))] \rangle / x \in X \},$$

$$A + B = \{ \langle x, [\inf M_A(x) + \inf M_B(x) - \inf M_A(x) \inf M_B(x), \sup M_A(x) + \sup M_B(x) - \sup M_A(x) \sup M_B(x)], [\inf N_A(x), \inf N_B(x), \sup N_A(x) \sup N_B(x)] \rangle / x \in X \},$$

$$A \cdot B = \{ \langle x, [\inf M_A(x) \inf M_B(x) - \sup M_A(x) \sup M_B(x), \sup M_A(x) \sup M_B(x)], [\inf N_A(x) + \inf N_B(x) - \inf N_A(x) \inf N_B(x), \sup N_A(x) + \sup N_B(x) - \sup N_A(x) \sup N_B(x)] \rangle / x \in X \},$$

$$A @ B = \{ \langle x, [1/2(\inf M_A(x) + \inf M_B(x)), 1/2(\sup M_A(x) + \sup M_B(x))], [1/2(\inf N_A(x) + \inf N_B(x)), 1/2(\sup N_A(x) + \sup N_B(x))] \rangle / x \in X \},$$

$$A \$ B = \{ \langle x = [\sqrt{\inf M_A(x), \inf M_B(x)}, \sqrt{\sup M_A(x), \sup M_B(x)}] \}$$

$$\{ \langle x = [\sqrt{\inf N_A(x), \inf N_B(x)}, \sqrt{\sup N_A(x), \sup N_B(x)}] \rangle / x \in X \}$$

$$A \# B = \{ \langle x, \left[\frac{2 \cdot (\inf M_A(x) \inf M_B(x))}{\inf M_A(x) + \inf M_B(x)}, \frac{2 \cdot (\sup M_A(x) \sup M_B(x))}{\sup M_A(x) + \sup M_B(x)} \right] \}$$

$$\left\{ \langle x, \left[\frac{2 \cdot (\inf N_A(x) \inf N_B(x))}{\inf N_A(x) + \inf N_B(x)}, \frac{2 \cdot (\sup N_A(x) \sup N_B(x))}{\sup N_A(x) + \sup N_B(x)} \right] \right\} / x \in X \}$$

$$A * B = \{ \langle x, \left[\frac{\inf M_A(x) + \inf M_B(x)}{2 \cdot (\inf M_A(x) \inf M_B(x) + 1)}, \frac{\sup M_A(x) + \sup M_B(x)}{2 \cdot (\sup M_A(x) \sup M_B(x) + 1)} \right] \}$$

$$\left\{ \langle x, \left[\frac{\inf N_A(x) + \inf N_B(x)}{2 \cdot (\inf N_A(x) \inf N_B(x) + 1)}, \frac{\sup N_A(x) + \sup N_B(x)}{2 \cdot (\sup N_A(x) \sup N_B(x) + 1)} \right] \right\} / x \in X \}$$

Let A be an IVIFS over E₁ and B over E₂, we define,

$$A \times_1 B = \{ \langle x, y \rangle, [\inf M_A(x), \inf M_B(y), \sup M_A(x) \sup M_B(y)], [\inf N_A(x), \inf N_B(y), \sup N_A(x) \sup N_B(y)] \rangle / x \in E_1, y \in E_2 \},$$

$$A \times_2 B = \{ \langle x, y \rangle, [\inf M_A(x) + \inf M_B(y) - \inf M_A(x) \inf M_B(y)], [\sup M_A(x) + \sup N_B(y) - \sup M_A(x) + \sup N_B(y)], [\inf N_A(x), \inf N_B(y), \sup N_A(x) \sup N_B(y)] \rangle / x \in E_1, y \in E_2 \},$$

$$A \times_3 B = \{ \langle x, y \rangle, [\inf M_A(x), \inf M_B(y), \sup M_A(x) \sup M_B(y)], [\inf N_A(x) + \inf N_B(y) - \inf N_A(x) \inf N_B(y), \sup N_A(x) + \sup N_B(y) - \sup N_A(x) \sup N_B(y)] \rangle / x \in E_1, y \in E_2 \},$$

$$A \times_4 B = \{ \langle x, y \rangle, \min(\inf M_A(x), \inf M_B(y), \sup M_A(x) \sup M_B(y)) \}$$

$$[\max(\inf N_A(x), \inf N_B(y)), \max(\sup N_A(x) \sup N_B(y))] / x \in E_1, y \in E_2 \},$$

$$A \times_5 B = \{ \langle x, y \rangle, [\max(\inf N_A(x), \inf N_B(y), \min(\sup N_A(x) \sup N_B(y)))] / x \in E_1, y \in E_2 \}.$$

It is easy to see the correctness of the defined operations and relations. The assertion can be transformed here with the necessary correction and their validity is changed similarly.

2.2. Interval valued Intuitionistic Fuzzy membership Rank correlation measure :

Let there be two non-empty intuitionistic fuzzy sets A and B defined on the same universe X. Let an element $x \in X$ belong to set A with membership value μ_A and non- membership value γ_A and to set B with membership value μ_B and non- membership value γ_B where μ and γ lie in the interval $[0,1]$ for all $x \in X$.

The proposed rank correlation measure should be symmetric with respect to μ, γ and it must also depend on the membership values of all $x \in X$.

Since conventional rank correlation measure lies in $[-1;1]$, it is expected that intuitionistic fuzzy measure should also lie in this range. The measure should also be independent of change of scale of membership and non-membership values. To achieve both the objectives we use a normalised form

Consider two intuitionistic fuzzy sets

First IFS is $[\mu_{AL}, \mu_{Au}, \gamma_{AL}, \gamma_{Au}]$

Second IFS is $[\mu_{BL}, \mu_{Bu}, \gamma_{BL}, \gamma_{Bu}]$

Where $\mu_{AL}, \mu_{BL}, \mu_{Au}, \mu_{Bu}$ are Membership

$\gamma_{AL}, \gamma_{BL}, \gamma_{Au}, \gamma_{Bu}$ are non_Membership each of the Variable

$[\mu_{AL}, \mu_{Au}, \gamma_{AL}, \gamma_{Au}] \& [\mu_{BL}, \mu_{Bu}, \gamma_{BL}, \gamma_{Bu}]$.

Takes the Values of $1, 2, 3, \dots, n$.

$$\begin{aligned} \text{Hence } [\mu_{AL} - \mu_{BL}] &= (\mu_{Au} - \mu_{Bu}) = (\gamma_{AL} - \gamma_{BL}) = (\gamma_{Au} - \gamma_{Bu}) = \left(\frac{n+1}{2}\right) \\ \sigma \mu_{L,u}^2 &= \frac{1}{n} \{ (\mu_{AL} - \mu_{BL}) - (\mu_{AL} - \mu_{BL}) + (\mu_{Au} - \mu_{Bu}) - (\mu_{Au} - \mu_{Bu}) \} \\ &= \frac{1}{n} \left\{ (1^2 + 2^2 + \dots + n^2 - \left(\frac{n+1}{2}\right)^2) + (1^2 + 2^2 + \dots + n^2 - \left(\frac{n+1}{2}\right)^2) \right\} \\ &= \frac{n^2-1}{12} + \frac{n^2-1}{12} \\ &= \frac{2(n^2-1)}{12} \end{aligned}$$

In general $d\mu_L = (\mu_{AL} - \mu_{BL}), d\mu_u = (\mu_{Au} - \mu_{Bu}), d\gamma_L = (\gamma_{AL} - \gamma_{BL}), d\gamma_u = (\gamma_{Au} - \gamma_{Bu})$

Let we consider,

$$d\mu_L = (\mu_{AL} - \mu_{BL}) - (\mu_{AL} - \mu_{BL})$$

$$d\mu_u = (\mu_{Au} - \mu_{Bu}) - (\mu_{Au} - \mu_{Bu})$$

$$d\gamma_L = (\gamma_{AL} - \gamma_{BL}) - (\gamma_{AL} - \gamma_{BL})$$

$$d\gamma_u = (\gamma_{Au} - \gamma_{Bu}) - (\gamma_{Au} - \gamma_{Bu})$$

Squaring and Subtracting

$$\begin{aligned} \{[\Sigma d\mu_L + \Sigma d\mu_u] - [\Sigma d\gamma_L + \Sigma d\gamma_u]\}^2 &= \{[(\mu_{AL} - \mu_{BL}) - (\mu_{AL} - \mu_{BL}) + (\mu_{Au} - \mu_{Bu}) - (\mu_{Au} - \mu_{Bu})] - \\ &[(\gamma_{AL} - \gamma_{BL}) - (\gamma_{AL} - \gamma_{BL}) + (\gamma_{Au} - \gamma_{Bu}) - (\gamma_{Au} - \gamma_{Bu})]\}^2 \\ &= [(\mu_{AL} - \mu_{BL}) - (\mu_{AL} - \mu_{BL}) + (\mu_{Au} - \mu_{Bu}) - (\mu_{Au} - \mu_{Bu})]^2 + [(\gamma_{AL} - \gamma_{BL}) - (\gamma_{AL} - \gamma_{BL}) + \\ &(\gamma_{Au} - \gamma_{Bu}) - (\gamma_{Au} - \gamma_{Bu})]^2 \end{aligned}$$

$$(\gamma_{Au} - \gamma_{Bu}) - (\gamma_{\square Au} - \gamma_{\square Bu})]^2 - 2[(\mu_{AL} - \mu_{BL}) - (\mu_{\square AL} - \mu_{\square BL}) + (\mu_{Au} - \mu_{Bu}) - (\mu_{\square Au} - \mu_{\square Bu})]$$

$$[(\gamma_{AL} - \gamma_{BL}) - (\gamma_{\square AL} - \gamma_{\square BL}) + (\gamma_{Au} - \gamma_{Bu}) - (\gamma_{\square Au} - \gamma_{\square Bu})]$$

$$= \Sigma \{ \sigma_{\mu_{LU}}^2 + \sigma_{\gamma_{u,L}}^2 - 2 \text{cov}(\mu_{LU}, \gamma_{LU}) \}$$

Let $\mu = \gamma$

$$\Sigma \{ \sigma_{\mu_{LU}}^2 + \sigma_{\gamma_{u,L}}^2 - 2\rho(\sigma_{\mu_{LU}} + \sigma_{\gamma_{u,L}}) \}$$

$$= \Sigma \{ \sigma_{\mu_{LL}}^2 + \sigma_{\mu_{LL}}^2 - 2\rho(\sigma_{\mu_{LL}} + \sigma_{\mu_{LL}}) \}$$

$$= \Sigma \{ 2\sigma_{\mu_{LL}}^2 - 2\rho\sigma_{\mu_{LL}}^2 \}$$

$$= 2\sigma_{\mu_{LL}}^2(1 - \rho)$$

Dividing by n

$$1/n \{ [\Sigma d\mu_L + \Sigma d\mu_U] - [\Sigma d\gamma_L + \Sigma d\gamma_U] \}^2 = 2\sigma_{\mu_{LL}}^2(1 - \rho)$$

$$(1 - \rho) = \frac{1}{n} \frac{[\Sigma[d\mu_L + d\mu_U] - \Sigma[d\gamma_L + d\gamma_U]]^2}{2\sigma_{\mu_{LL}}^2}$$

$$\rho = 1 - \frac{[\Sigma[d\mu_L + d\mu_U] - \Sigma[d\gamma_L + d\gamma_U]]^2}{2n\sigma_{\mu_{LL}}^2}$$

$$\rho = 1 - \frac{6 \{ [\Sigma[d\mu_L + d\mu_U] - \Sigma[d\gamma_L + d\gamma_U]]^2 \}}{2n(n^2 - 1)}$$

Problem;

To find the Rank correlation between two Interval valued Intuitionistic fuzzy set.

SL.NO	A				B			
	μ_{AL}	μ_{BL}	μ_{AU}	μ_{BU}	γ_{AL}	γ_{BL}	γ_{AU}	γ_{BU}
1	0.005	0.006	0.004	0.006	0.001	0.007	0.008	0.018
2	0.003	0.005	0.010	0.012	0.022	0.024	0.024	0.025
3	0.006	0.007	0.012	0.016	0.018	0.020	0.022	0.023
4	0.005	0.007	0.014	0.018	0.020	0.024	0.025	0.026
5	0.001	0.004	0.020	0.022	0.024	0.026	0.028	0.030
6	0.002	0.003	0.022	0.024	0.024	0.028	0.001	0.009
7	0.004	0.005	0.018	0.020	0.026	0.030	0.008	0.010
8	0.001	0.002	0.012	0.015	0.012	0.018	0.009	0.015
9	0.004	0.007	0.011	0.012	0.017	0.019	0.010	0.015
10	0.008	0.010	0.014	0.018	0.006	0.008	0.015	0.018

SOLUTION

$$\rho = 1 - \frac{6 \{ [\Sigma[d\mu_L + d\mu_U] - \Sigma[d\gamma_L + d\gamma_U]]^2 \}}{2n(n^2 - 1)}$$

$$d\mu_L=(\mu_{AL}-\mu_{BL}), d\mu_u=(\mu_{Au}-\mu_{Bu}), d\gamma_L=(\gamma_{AL}-\gamma_{BL}), d\gamma_u=(\gamma_{Au}-\gamma_{Bu})$$

$d\mu_L=(\mu_{AL}-\mu_{BL})$	$d\mu_u=(\mu_{Au}-\mu_{Bu})$	$d\gamma_L=(\gamma_{AL}-\gamma_{BL})$	$d\gamma_u=(\gamma_{Au}-\gamma_{Bu})$	$d\mu_L+d\mu_u$	$d\gamma_L+d\gamma_u$	$(d\mu_L+d\mu_u)+(d\gamma_L+d\gamma_u)$	$\{(d\mu_L+d\mu_u)+(d\gamma_L+d\gamma_u)\}^2$
-0.001	-0.002	-0.006	-0.010	-0.003	-0.016	-0.019	0.000361
-0.002	-0.002	-0.002	-0.001	-0.004	-0.003	-0.009	0.000081
-0.001	-0.004	-0.002	-0.001	-0.005	-0.003	-0.008	0.000064
-0.002	-0.004	-0.004	-0.001	-0.006	-0.005	-0.011	0.000121
-0.003	-0.002	-0.002	-0.002	-0.005	-0.004	-0.009	0.000081
-0.001	-0.002	-0.004	-0.008	-0.003	-0.012	-0.015	0.000225
-0.001	-0.002	-0.004	-0.002	-0.003	-0.006	-0.009	0.000081
-0.001	-0.003	-0.006	-0.006	-0.004	-0.012	-0.016	0.000256
-0.004	-0.001	-0.002	-0.005	-0.005	-0.007	-0.012	0.000144
-0.002	-0.004	-0.002	-0.003	-0.006	-0.005	-0.011	0.000121

$$\rho = 1 - \frac{6 \{ \sum [d\mu_L + d\mu_U] - \sum [d\gamma_L + d\gamma_U] \}^2}{2n(n^2 - 1)}$$

$\sum d\mu_L = -0.018, \sum d\mu_u = -0.026, \sum d\gamma_L = -0.034, \sum d\gamma_u = -0.039, n=10$

$$\rho = 1 - \frac{6 \{ [(-0.018) + (-0.026)] - [(-0.034) + (-0.039)] \}^2}{2(10)(10^2 - 1)}$$

$$= 1 - \frac{6 \{ [-0.044] - [-0.0730] \}^2}{2(10)(10^2 - 1)}$$

$$= 1 - \frac{6 \{ (-0.117) \}^2}{(20)(100 - 1)}$$

$$= 1 - 0.7020$$

$$= 1 - \frac{(-0.7020)}{1980}$$

$$= 1 - 0.0003545$$

$\rho = 0.99$

Conclusion:

Already create fuzzy rank correlation in [3] and we have used the fuzzy rank correlation coefficient for creating intuitionistic fuzzy rank correlation. And also we derived the numerical example in intuitionistic fuzzy rank correlation.

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